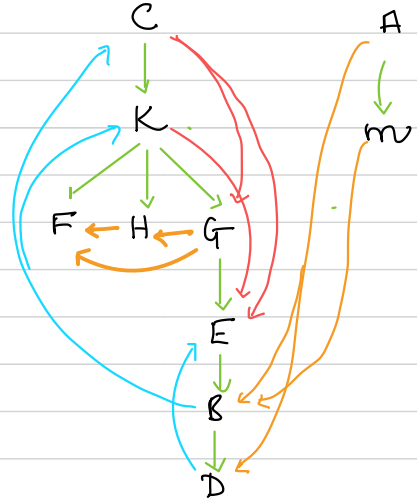
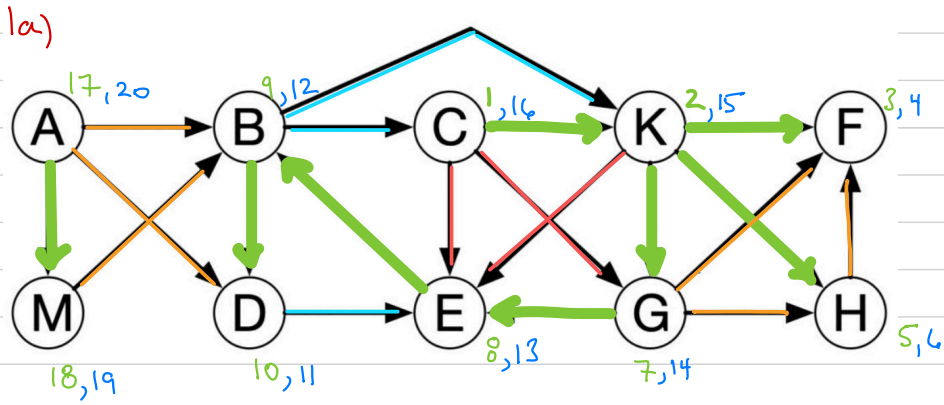
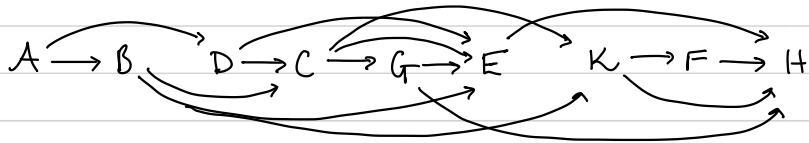
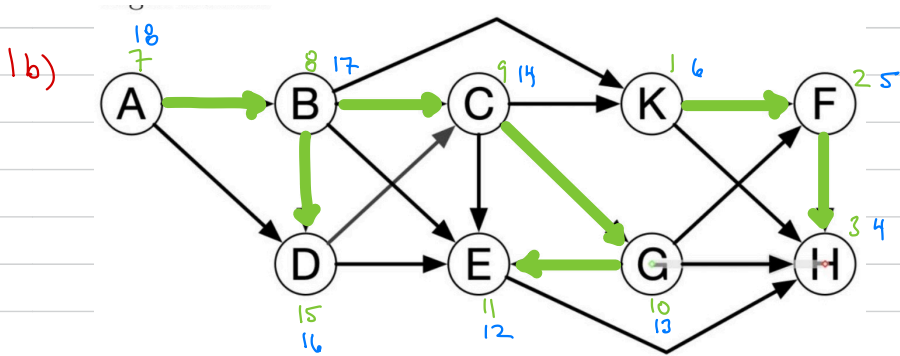


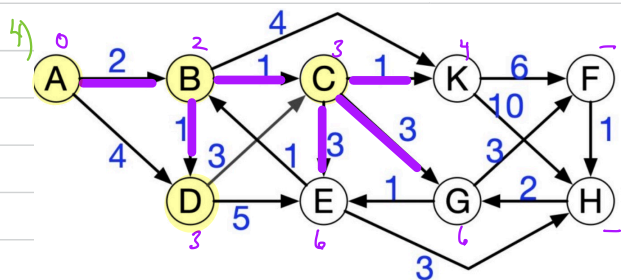
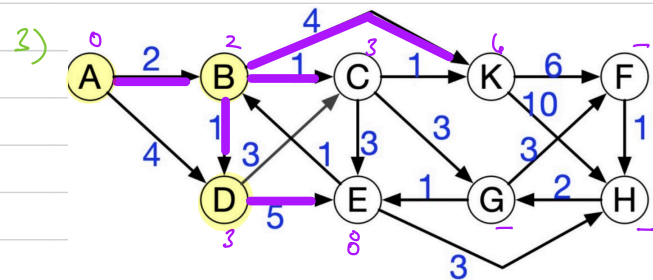
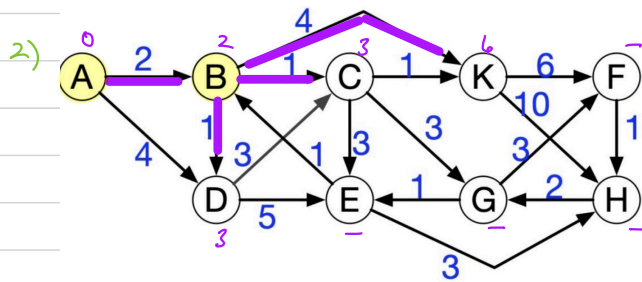
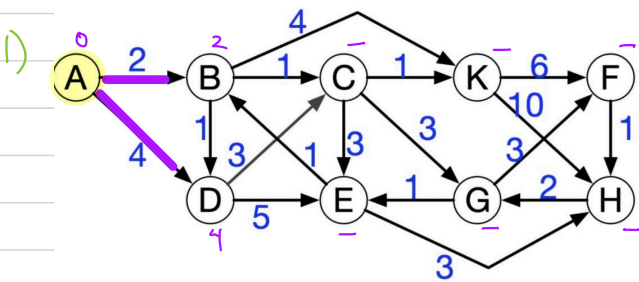
Assignment 5 solutions



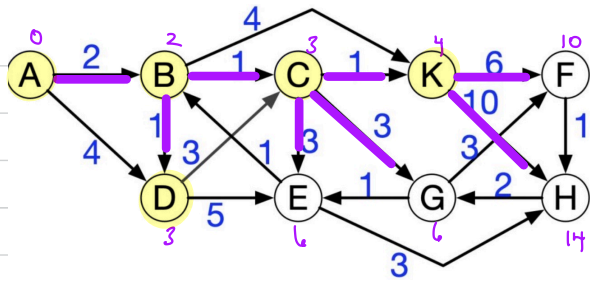
— Forward — Back — Back



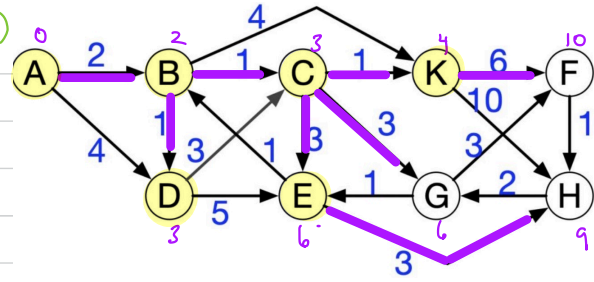
1c) yellow vertices have been removed from G



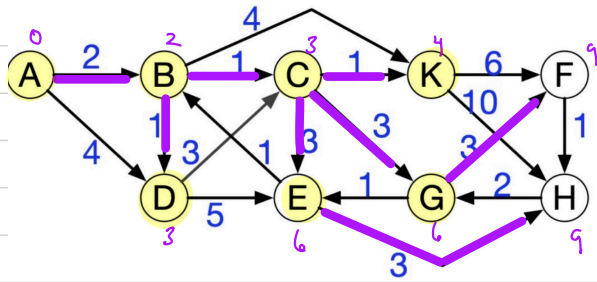
5)



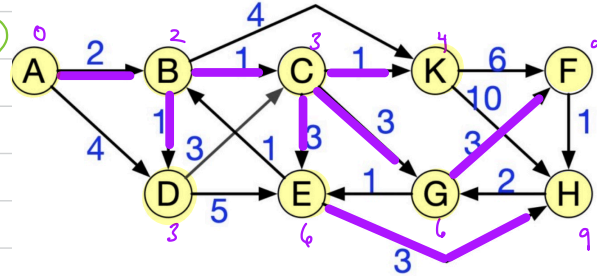
6)



7)

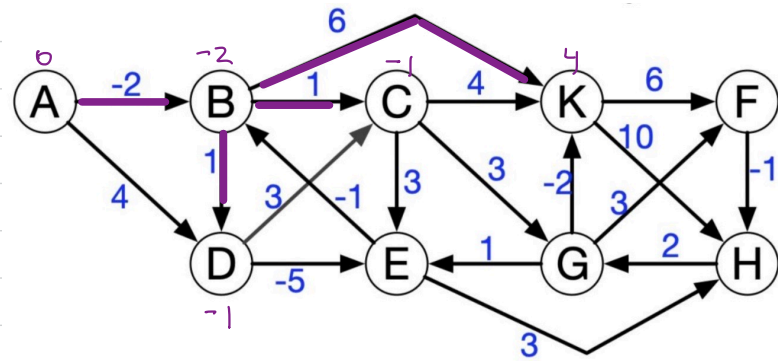
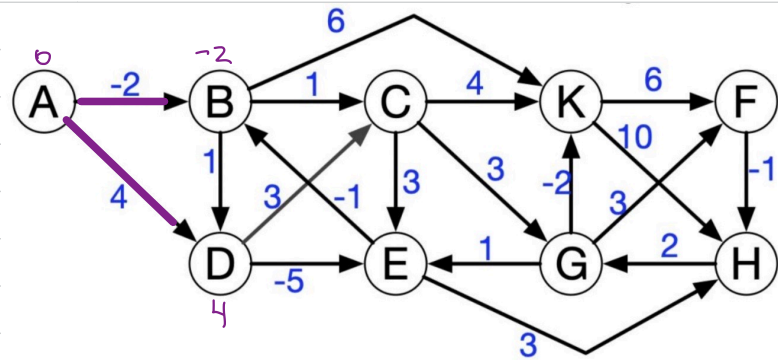


8)



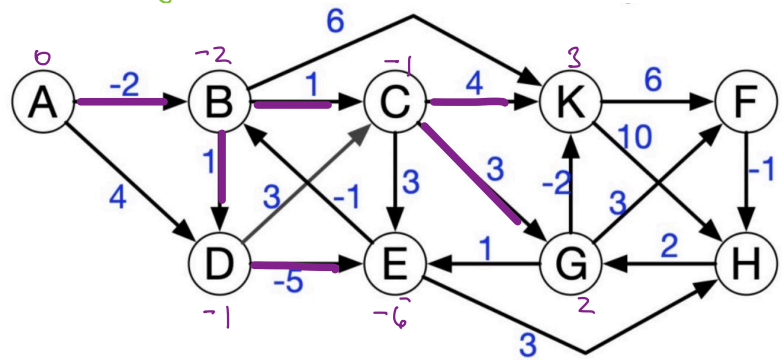
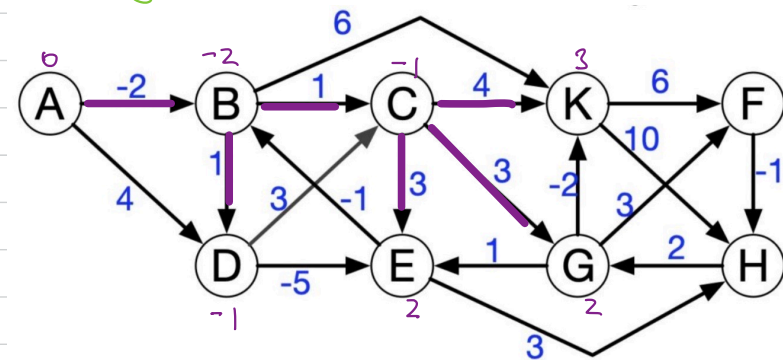
(e) 1) Edges out of A:

2) Edges out of B:



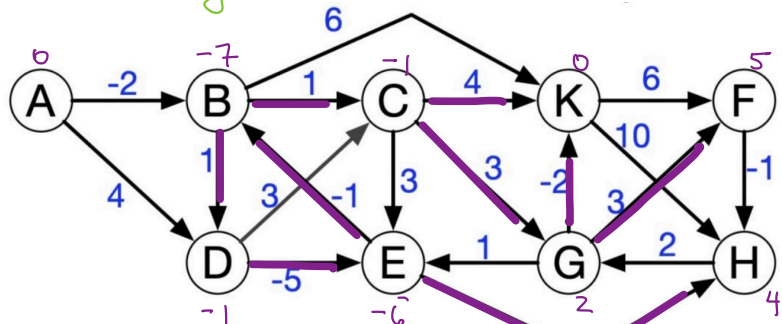
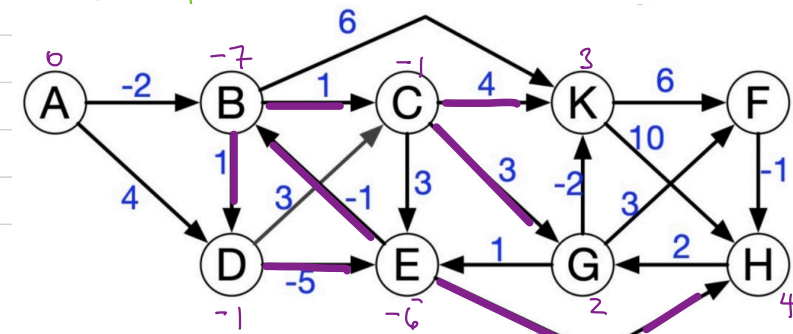
3) Edges out of C:

4) Edges out of D:



5) Edges out of E:

6) Edges out of F, G:



7) Edges out of H, K : no change // End of iteration 1.

For each subsequent iteration, B.d will decrease by 5. Therefore at iteration 9 a neg weight cycle will be detected.

2a) Update DFS visit so that it returns the max. cycle length down the DFS tree from vertex u . Each time a NEW cycle is found, we compare its length to the current max.

Since this is a simple (constant-time) change to DFS, the runtime is $O(V+E)$.

Initialise: Start at any vertex s , set $s.dist = 0$

MaxCycle(u)

maxc = 0

$u.visited = TRUE$

for all v in $Adj(u)$

if $v.visited = False$

$v.parent = u$

$v.dist = u.dist + 1$

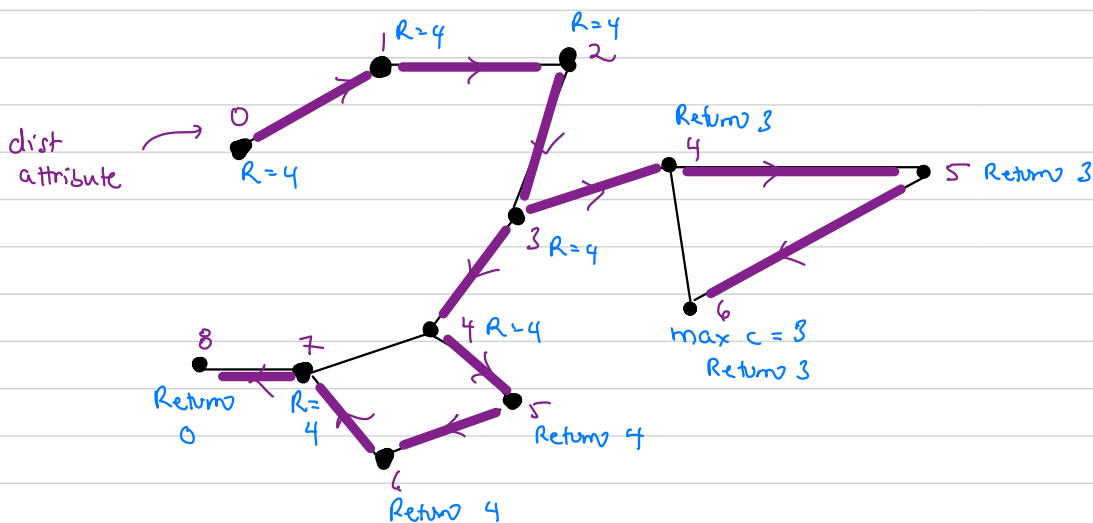
$maxc = \max(maxc, \text{MaxCycle}(v))$ // update maxc

else if $v \neq u.parent$ // found backedge = cycle

$L = u.dist - v.dist + 1$ // new cycle length

$maxc = \max(L, maxc)$

Return maxc.



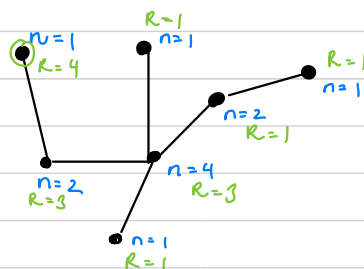
- b) Start the procedure below using any vertex u from V . The constant-time change to DFS means that the runtime is $O(V+E)$

CountLeaves(u)

```

u.visited = TRUE
count = 0
Neighbors = {}
for all  $v$  in Adj( $u$ )
    Neighbors +=  $v$ 
    if  $v$ .visited = False
         $v$ .parent =  $u$ 
        count += CountLeaves( $v$ )
if Neighbors = {}
    Return 1 + count
else Return count

```



- c) Model the graph as a directed unweighted graph, where vertex = farm, and directed edge is one way road. Additionally, each vertex has attribute $v.scc$ to identify which component it's in.

Step 1) Run DFS(G) on original graph, where DFS-visit adds terminated vertices to the back of list L .

DFS-visit(u)

```

u.visited = TRUE
for all  $v$  in Adj( $u$ )
    if  $v$ .visited = False
        DFS-visit( $v$ )

```

$L.add(u)$

Now L is a list of vertices in decreasing order of finish time.

Runtime: $O(V+E)$

Step 2) Create new graph G^T using edge set E^T which contains the original edges from E but in reverse order. $O(E)$ for reversing edges.

Step 3) Use first vertex from L and run DFS-visit(u) with the following update:

DFS-visit(u)

```

u.visited = TRUE
u.scc = 1
for all  $v$  in Adj( $u$ )
    if  $v$ .visited = False
        DFS-visit( $v$ )

```

$O(V+E)$

Now each vertex in SCC 1 is indicated by attribute $u.scc = 1$.

Step 4) Find another vertex that is part of the 2nd SCC by finding an unvisited vertex.

for all v in V

if v .visited = False

$s = v$

break

$O(V)$

Step 5) Loop through the vertices of SCC1 and identify a vertex that can be connected to s . Note that since s is in SCC2, we know $SCC1 \rightarrow SCC2$, but not the reverse.

```

for all  $v$  in  $V$ 
    if  $v.scc = 1$ 
        if  $(v, s) \notin E$ 
            Print add edge  $(s, v)$  to  $G$ .
            break
    }  $O(V)$ 

```

Runtime: $O(V+E) = O(n+n^2) = O(n^2)$ since $|E|$ is

(d) Step 1) Run DFS from S , using only vertices for which $b(u, v) = \text{false}$.
Use visited attribute $v.fromS$

DFS-visit(u)

$u.fromS$

for all v in $Adj(u)$

if $b(u, v) = \text{false}$ // not a bridge

if $v.fromS = \text{false}$

$v.parent = u$

DFS-visit(v)

Step 2) Now all visited vertices are those we can reach from S without using a bridge. If $T.fromS = T$, return true.

Step 3) Run DFS from vertex T , again using only edges $b(u, v) = \text{false}$.
Use visited attribute $v.fromT$

Step 4) Find if there is 1 bridge connected each path:

for all $e = (u, v)$ in E

if $b(u, v) = \text{TRUE}$ // bridge

if $u.fromS = T$ and $v.fromT = T$

return TRUE

Return False

Runtime: Each DFS runs in $O(V+E)$ and step 4 in time $O(E)$.
∴ total runtime $O(V+E)$.

(e) Update step 2 above: if $T.fromS = T$ then call PrintBackwards(T)
else ... in step 4:

for all $e = (u, v)$ in E

if $b(u, v) = \text{TRUE}$

if $u.fromS = T$ and $v.fromT = T$

PrintBackwards(u)

PrintForwards(v)

Return TRUE

Print Backwards (u)

if $u \neq w$

Print Backwards (u.parent)

Print u

PrintForwards (u)

if $u \neq \text{NIL}$

Print u

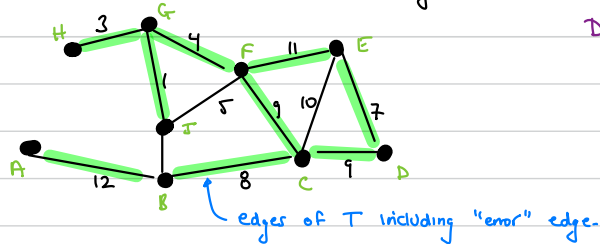
```
PrintForwards(u.parent)
```

3a)

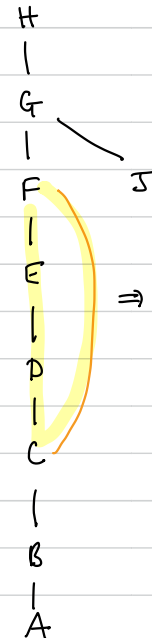
Create a new graph G' with vertex set V and edges the set of edges in T (including the error). $|T| = O(V)$.

Note that the "error" edge must have created a cycle. We can
 run Find Cycle on G' to find the back edge which is part of the cycle.
 This runs in time $O(V+E) = O(V+V) = O(V)$.

Once we find the backedge, we can follow the parent references along the edges of the cycle, until we hit the other end of the backedge. These edges along with the back edge form a cycle. We can take the max. weight out of these edges. This represents the "bad" edge.



DFS:



⇒ Cycle. Biggest edge weight is 11,
which is NOT part of MST.
∴ remove edge FE.

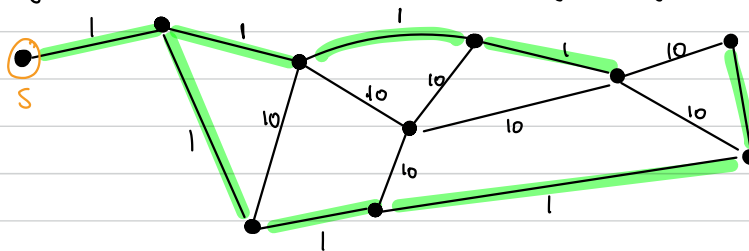
(b) If all edge weights are distinct, only 1 MST is possible.

Proof: Suppose (for contradiction) that there was more than

one MST: T_1 and T_2 , each of equal (minimum) weight. Then some edge e_1 in T_1 is NOT in T_2 .

Similarly $\exists e_2 \in T_2$ that is not in T_1 . Suppose $e_1 < e_2$. If we add edge e_1 to T_2 we create a cycle. The largest edge on this cycle is not in the MST, but making this edge swap means creating a smaller MST. Which is impossible!

(c)


$$\text{Prims}(S) = \text{Dijkstra}(S)$$

4a)

Model as a graph with vertices as intersections and undirected edges as the bidirectional roads. The edge weights are the toll values, and u-tax is a vertex attribute.

Use a variation of Dijkstra's algorithm:

Initialise: Initialise $v.d = \infty$ for all vertices.
 Set $s.d = s.tax$
 Insert all vertices into min Priority Queue.

Main Loop: While Q not empty:
 $u = \text{Extract_min}(Q)$
 for all v in $\text{Adj}(u)$ where $v \in Q$
 if $v.d > u.d + w(u,v) + v.tax$
 $\text{DecreaseKey}(Q, v, u.d + w(u,v) + v.tax)$
 $v.parent = u$

Return values: Print $t.d$
 PrintBackwards(t)

Runtime: Dijkstra runs in $O(E \log V)$
 $= O(m \log n)$

PrintBackwards(u)

IF $u \neq \text{NIL}$
 PrintBackwards($u.parent$)
 Print u .

4b) Let the 5 trail markers with toilets T_1, T_2, T_3, T_4, T_5 .

Step 1) Run Dijkstra's algorithm from S , storing distances in $v.fromS$
 Any vertices for which $v.fromS > 20$, reset $v.fromS = \infty$.

Step 2) Run Dijkstra's algorithm from T_1 , storing distance in $v.fromT_1$
 Any vertices for which $v.fromT_1 > 20$, reset $v.fromT_1 = \infty$
 Repeat for T_2, T_3, T_4, T_5 .

Step 4) Create a new graph with vertex set $S, F, T_1, T_2, T_3, T_4, T_5$.
 The edges are directed and weighted. The weight of each edge corresponds to an attribute from the original graph:

$$W(S, T_1) = T_1.fromS$$

$$W(S, T_2) = T_2.fromS$$

\vdots

$$W(T_2, T_1) = T_1.fromT_2$$

} Set all edge weights. Recall some edge weights will be ∞ .

Step 5) Run Dijkstra's from S on the new graph, storing distances in $v.d$. The value in $F.d$ represents the shortest distance from $S \rightarrow F$ while respecting the conditions.

Runtime: Each call to Dijkstra runs in $O(E \log V) = O(m \log n)$.

The last call to Dijkstra runs in constant time since the graph has a constant # of vertices.

\therefore overall runtime is $O(m \log n)$